

#### Bridging the Dimensions: Seamless Integration of 3D Structure-based Design and 2D Structure-activity Relationships to Guide Medicinal Chemistry

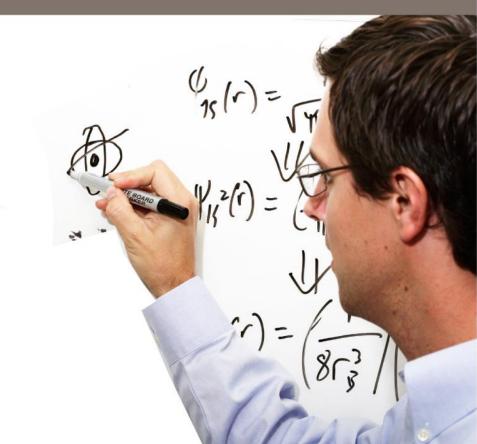
ACS Spring National Meeting. COMP, March 13<sup>th</sup> 2016 Marcus Gastreich, <u>Matthew Segall</u>, Carsten Detering, Ed Champness, Christian Lemmen matt.segall@optibrium.com

## **Overview**

- 2-dimensional (2D) structure-activity relationships (SAR)
  - Qualitative: Activity cliffs, matched molecular pair analysis...
  - Quantitative: QSAR models
- 3-dimensional (3D) structure-based design
  - Scoring/affinity prediction
  - Understanding 3D SAR
- Linking 2D and 3D SAR to guide design
- Conclusions

## 2D Structure-Activity Relationships





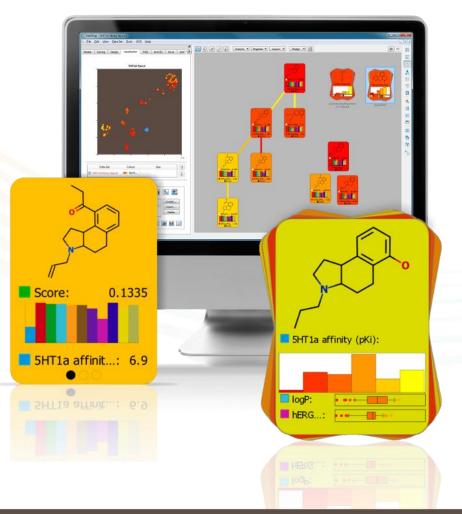
## **Qualitative SAR**

Many methods are routinely used for analysis of data to reveal patterns and trends to guide compound optimisation, e.g.

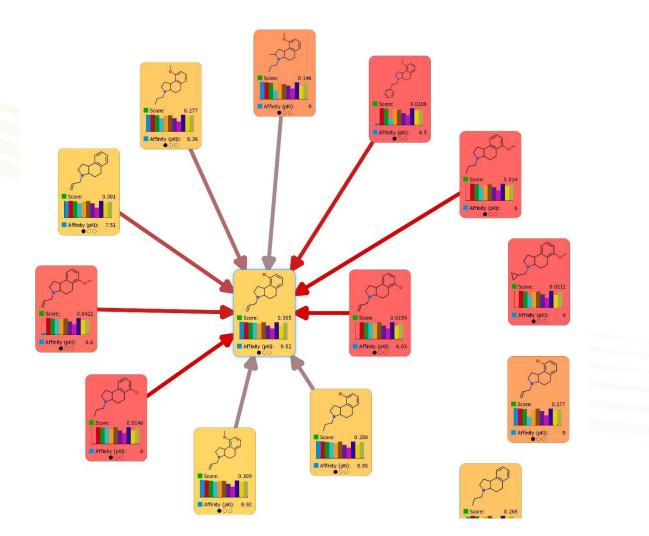
- Clustering
  - Group 'similar' compounds to identify series with interesting SAR
- Activity cliff detection
  - Small changes in structure that cause a large change in activity
- Matched molecular pair analysis
  - Pairs of compounds that are identical except for one small change at the same position

### Visualising 2D SAR Card View™

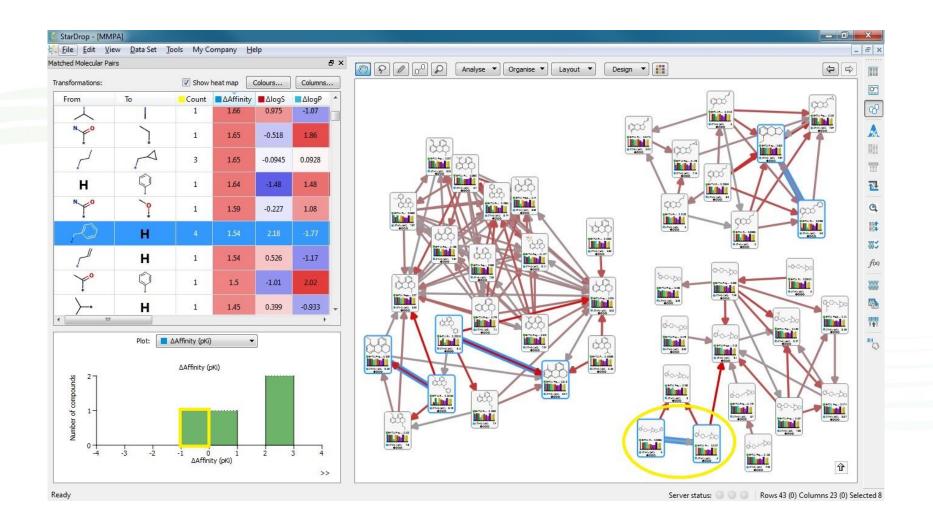
- Freedom from the constraints of 'chemical spreadsheets'
  - Represent compound relationships
- Work the way you think
  - Cards: Display key compound data
  - Stacks: Summarise and compare data for groups of compounds
  - Links: Highlight compound relationships
- Intuitive visualisation of analyses
  - Clustering, activity cliffs, matched molecular pairs...
- Quickly identify optimisation strategies



#### Activity Neighbourhood Activity Cliff Visualisation



© 2016 Optibrium Ltd. M.D. Segall *et al.* (2015) Drug Discov. Today **20**(9) pp. 1093-1103



### Quantitative Structure-Activity Relationships Principles

 $y = f(x_1, x_2, x_3, ...) \pm \varepsilon$ 

• Data

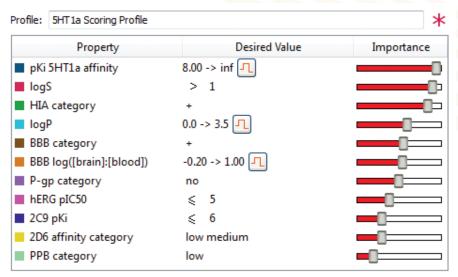
Statistical uncertainty

- Quality data is essential
- Public data needs very careful curation (and may not be good enough)
- Descriptors, e.g.
  - Whole molecule properties, e.g. logP, MW, PSA...
  - Structural descriptors, SMARTS, fingerprints...
- Statistical fitting or machine learning method, e.g.
  - Partial least squares, artificial neural networks, support vector machines, random forest, Gaussian processes...
- Widely applied to prediction of ADME and physicochemical properties

#### Multi-Parameter Optimisation Probabilistic Scoring

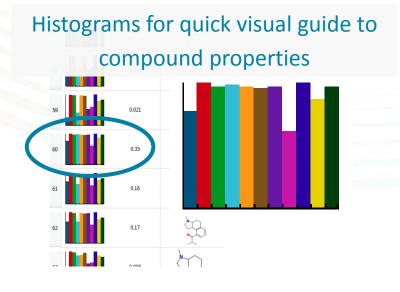
#### Integrated assessment of data against project criteria

Accounts for the uncertainties in all compound-related data (experimental or calculated)



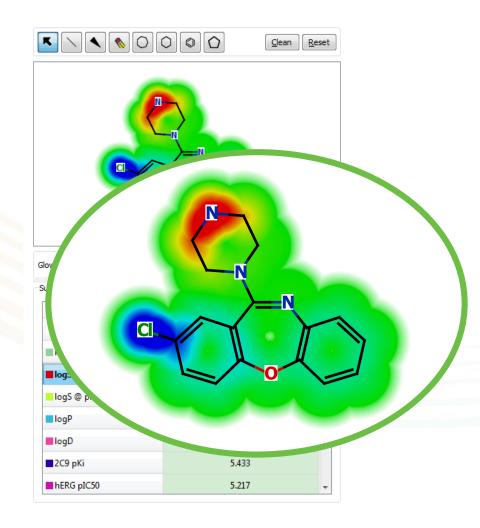
#### Project specific scoring profile

Compounds ranked by likelihood of success



## **Interactive Redesign**

- QSAR models provide estimates of compounds' properties
- Instant feedback on how properties are likely to change
  - Explore strategies for redesign
- But, important questions
  - "Why is a property value predicted?"
  - "Where can I change this property?"
- Glowing Molecule<sup>™</sup>:
  - Visual indication of structural influences on predicted properties

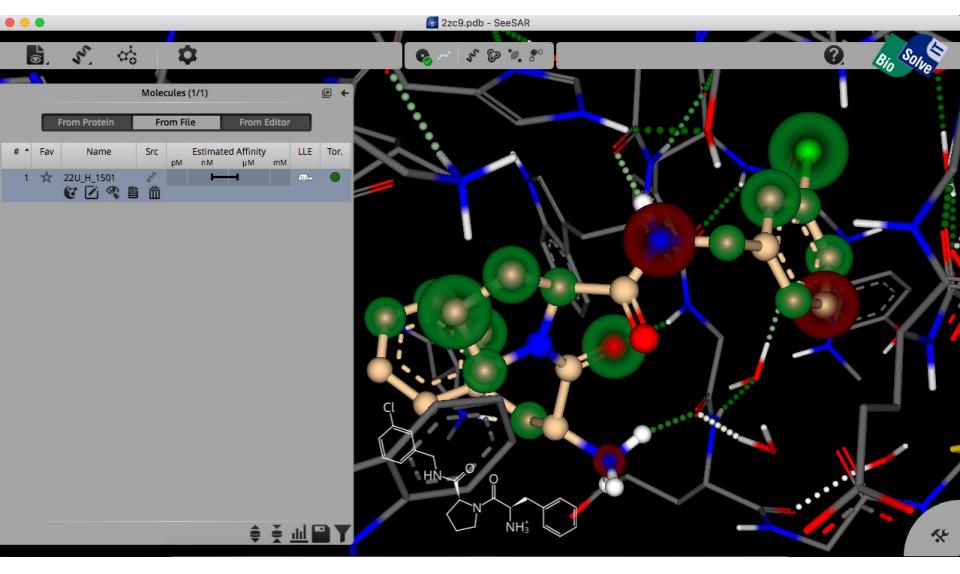




# **3D Structure-Based Design**



# Visual Understanding of 3D Affinity Data





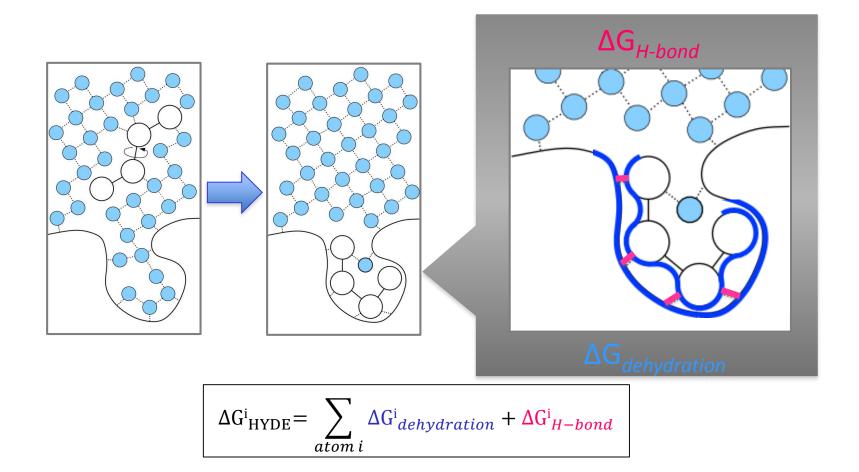
## HYDE: A Different View @ Energetics\*

- Physics only
- No calibration to complexes
- Relates to real Free Energies



\*CINF 9: Christian Lemmen, "Predicting binding affinity doesn't work, or does it?" Next talk!

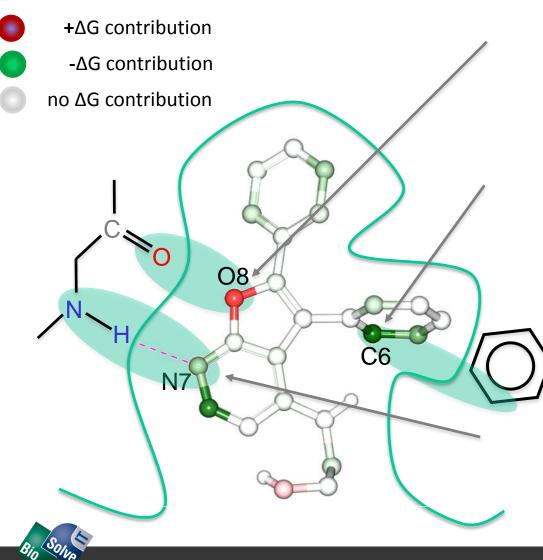
# **HYDE Scoring Function – Concept\***





# **Hyde - Visual Affinities**

#### HYDE color code:

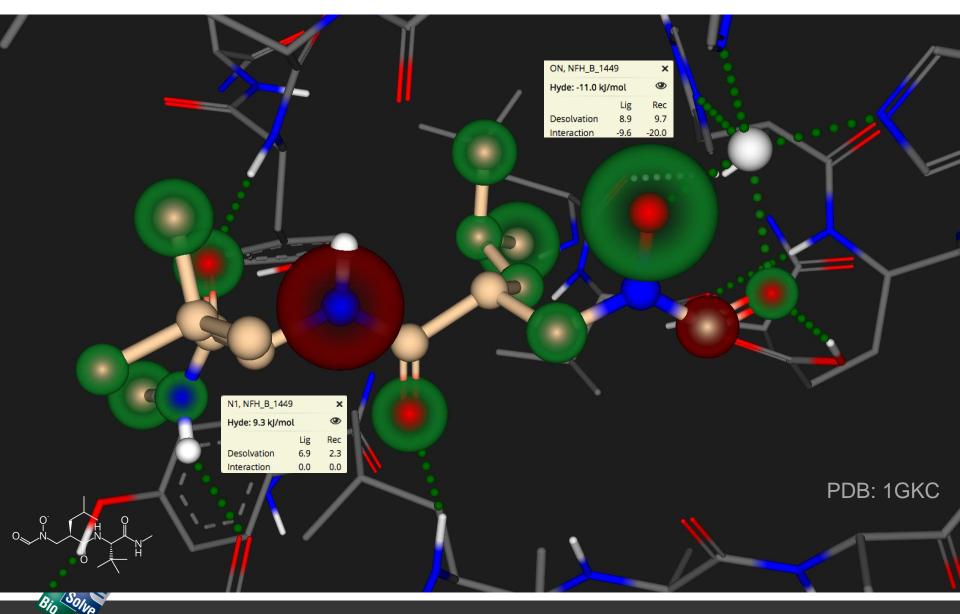


total desolvation cost	<u>10.6</u> kJ/mol
ligand aromatic oxygen	2.4 kJ/mol
receptor carbonyl oxygen	8.2 kJ/mol

receptor aromatic carbons	-5.2 kJ/mol
ligand aromatic carbon	-2.0 kJ/mol
total desolvation gain	<u>-7.2</u> kJ/mol

receptor amide N dehydrat	6.3	kJ/mol
interaction energy	-7.4	kJ/mol
ligand aromatic N dehydrat	6.4	kJ/mol
interaction energy	-7.5	kJ/mol
total H-bond energy	<u>-2.2</u>	kJ/mol

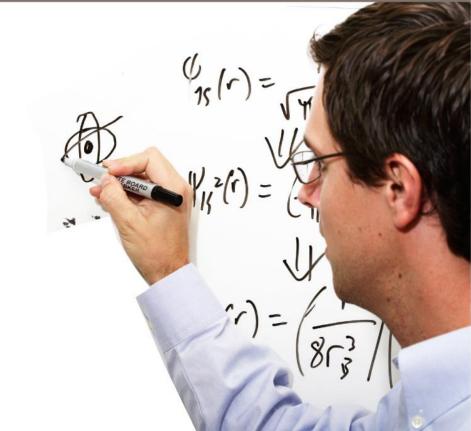
## **Visual Understanding of 3D Affinity Data**

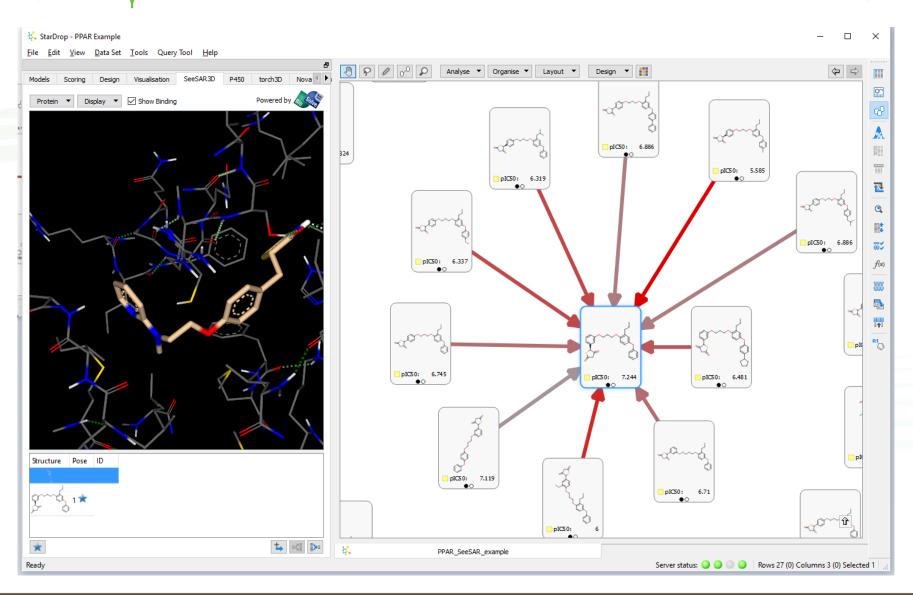


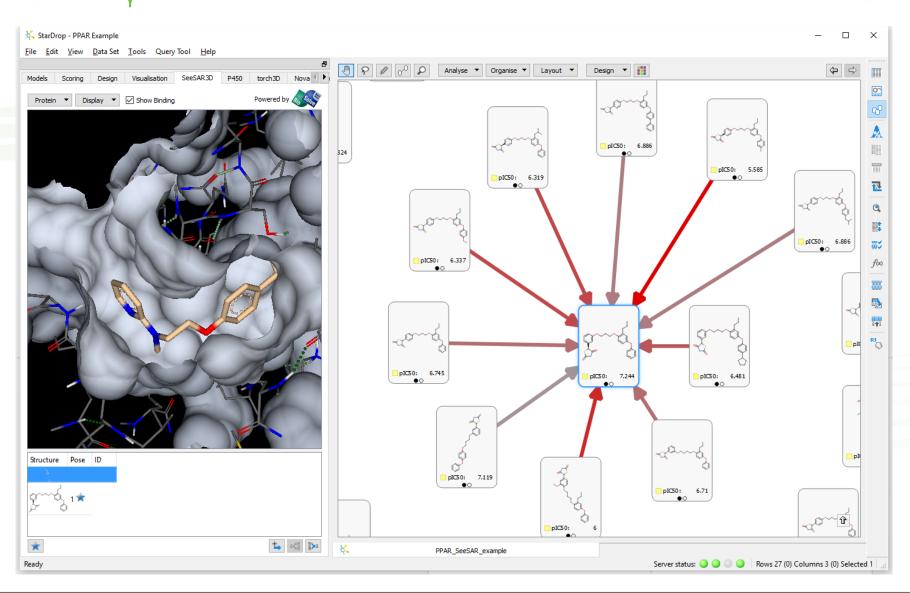
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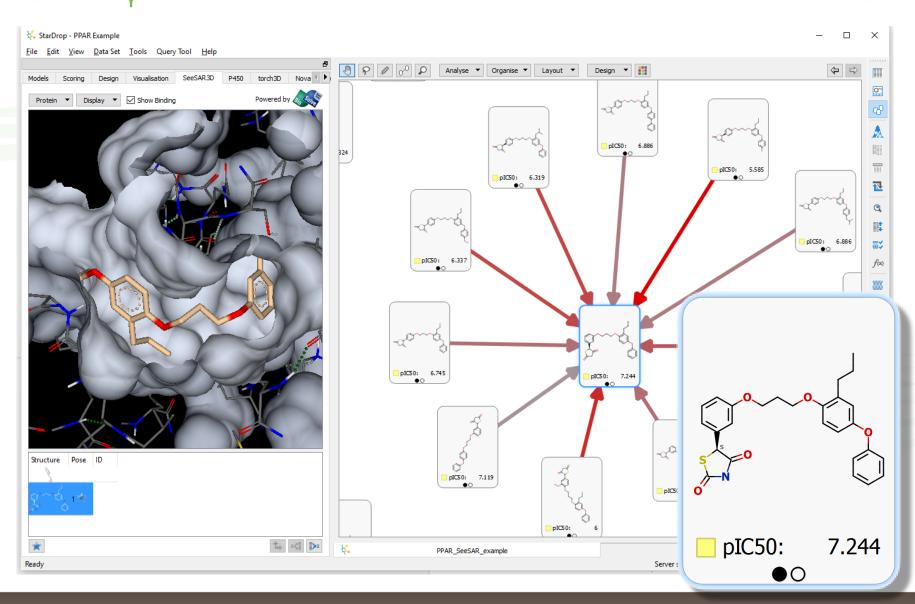
## Linking 2D and 3D SAR to Guide Design

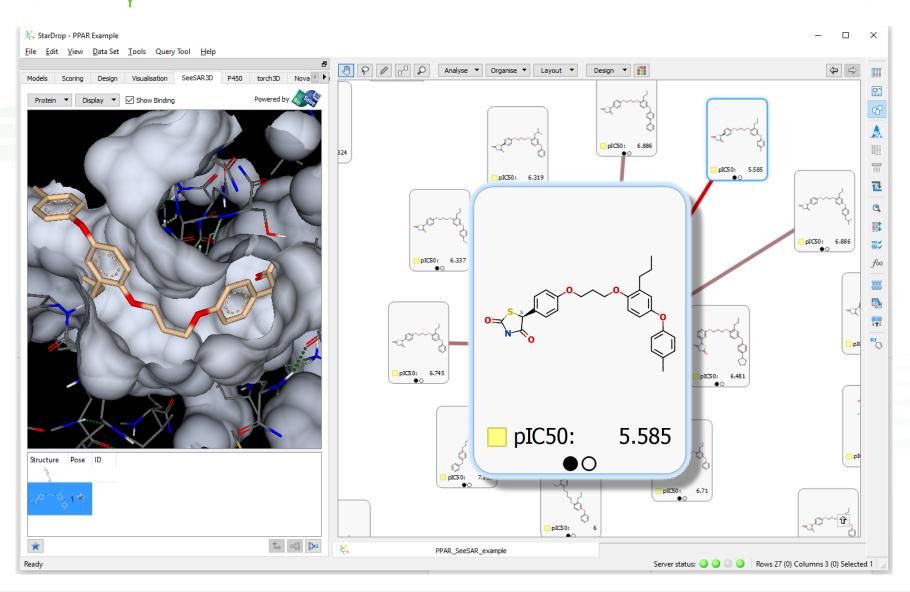


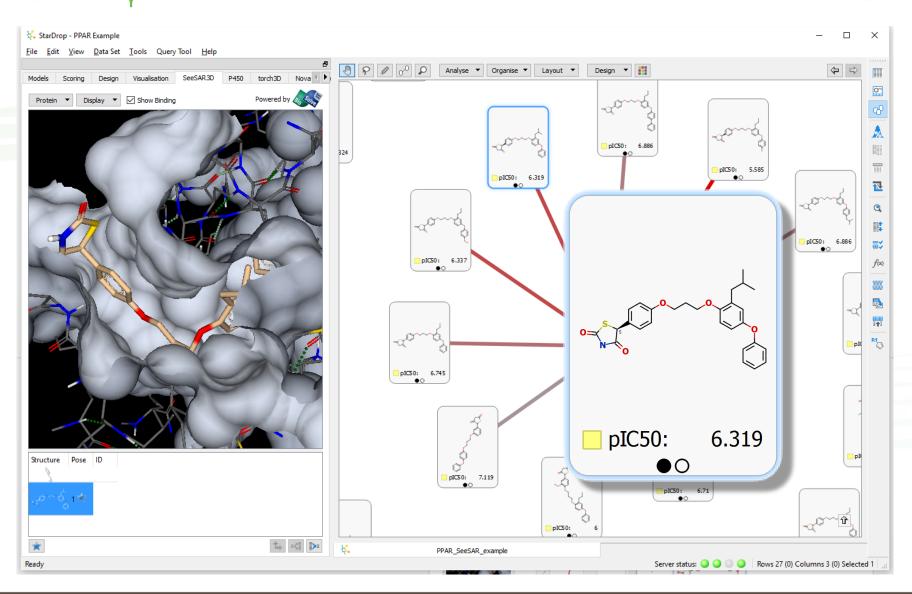






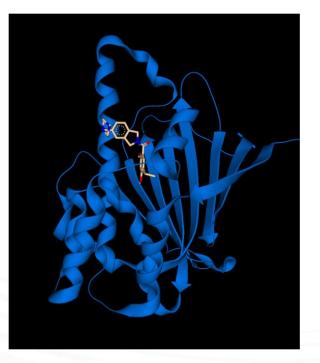


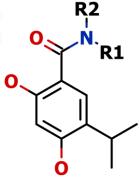


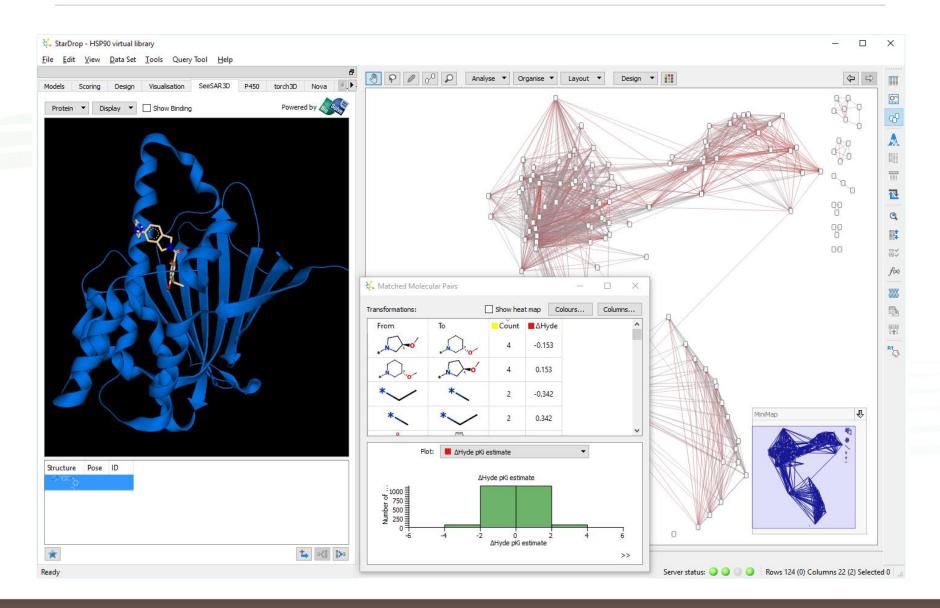


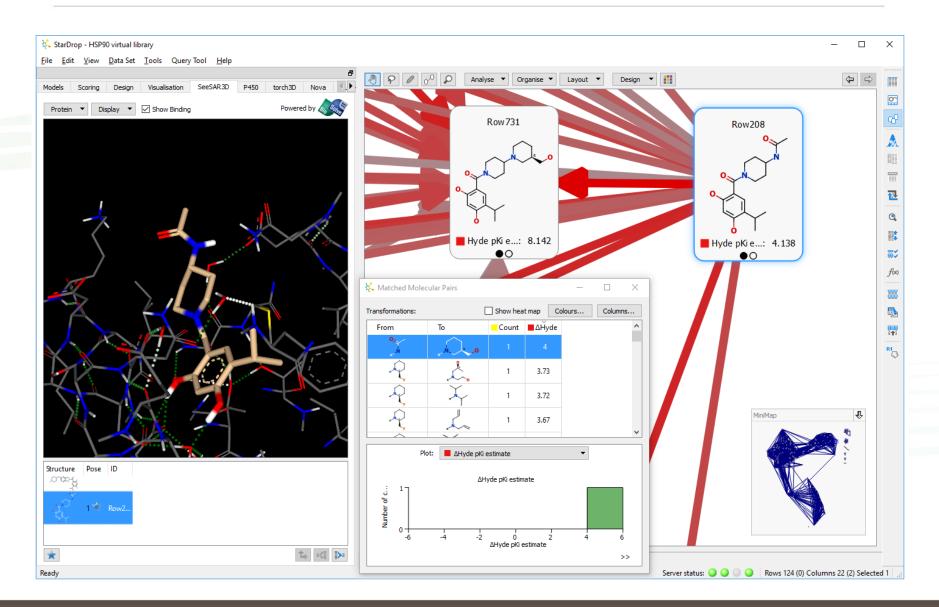
#### Exploration of Virtual Screening Results HSP90

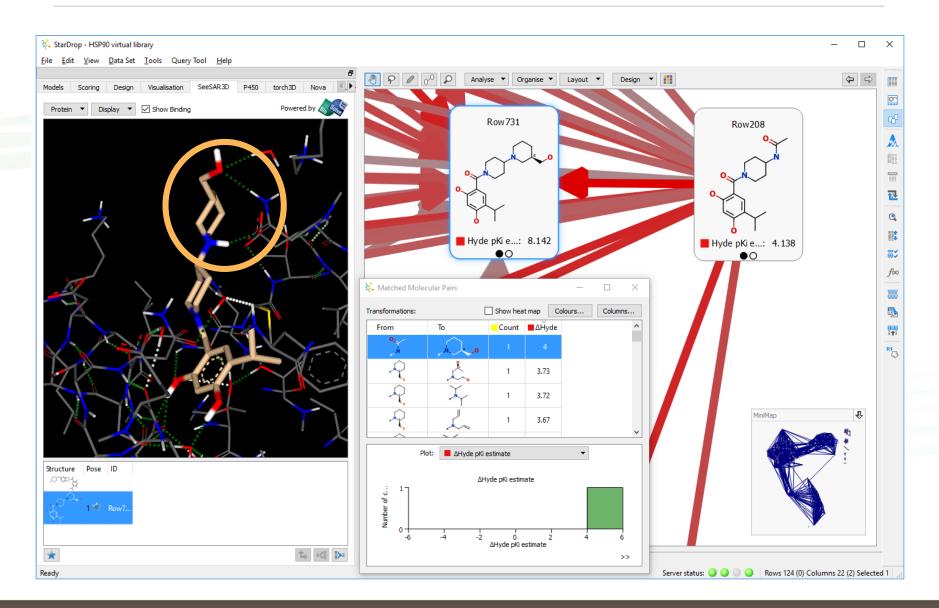
- Crystal structure PDB ref. 2XJX
- Virtual library generated using STORM workflow in KNIME
  - Amide substitution using Schotten
     Baumann reaction on beta resorcylic core
  - Building blocks from vendor catalogues
  - 'Tail' of molecule not contributing to affinity
- Resulting library docked with FlexX
- Scored using SeeSAR and HYDE to estimate pK<sub>i</sub>

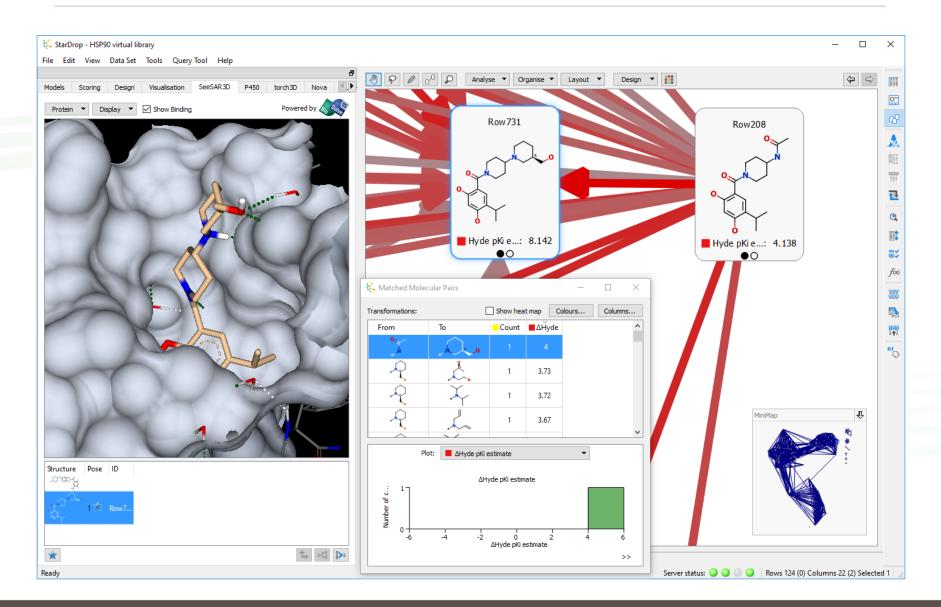




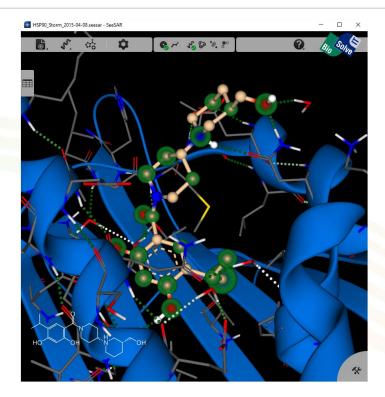


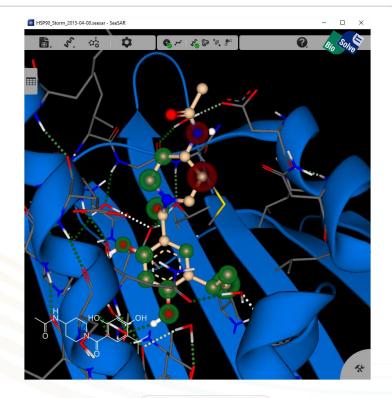


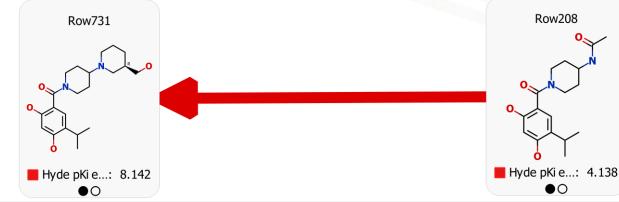




## **HYDE Analysis in SeeSAR**



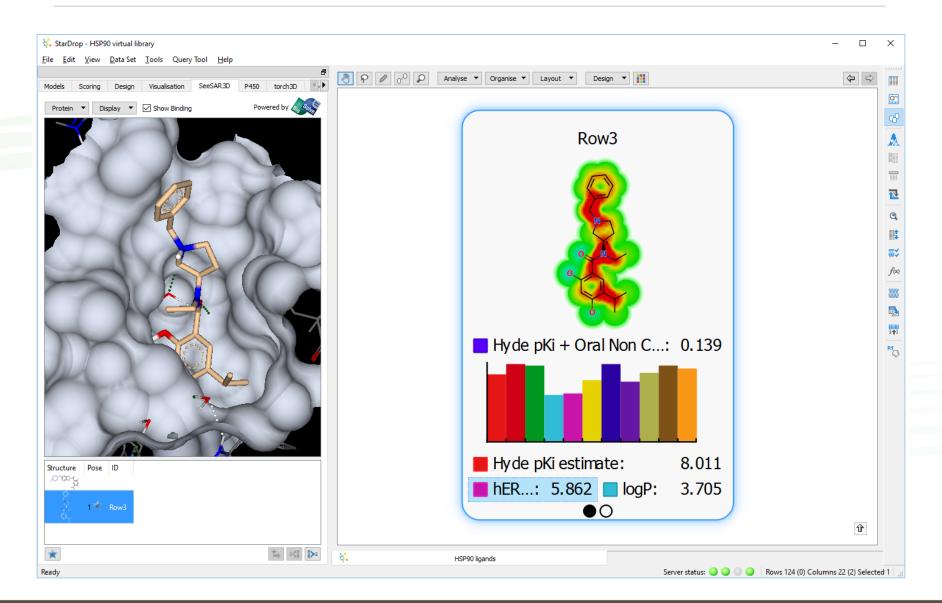




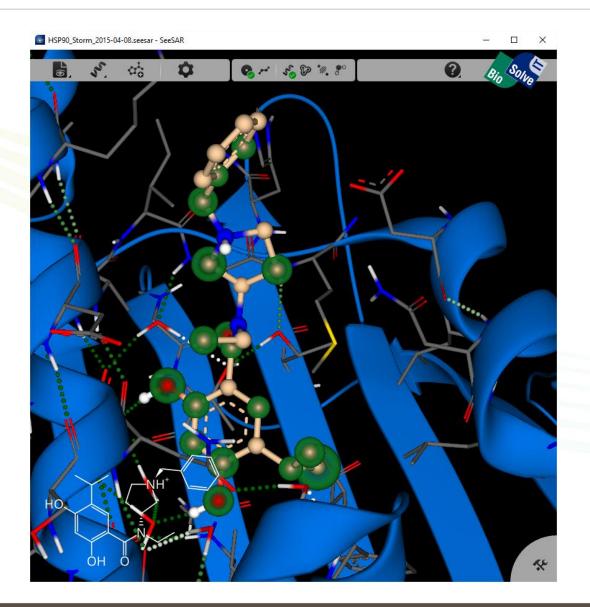
#### Combine with 2D QSAR Predictions Multi-parameter optimisation

dels Scoring Design Visualisation SeeSAR3D P450 torch3D	₽	Hyde pKi	+ Oral Non CN	Structure	D	Hyde pKi	logS	logP	hERG pIC5	BBB log([l	2C9 pKi	^
file: Hyde pKi + Oral Non CNS Scoring Profile		<b>Unite</b>		Ì	1							
Property Desired Value Importance	= 1		0.4134	La	Row36	7.52	4.185	1.53	4.55	-0.7597	4.301	
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logS > 1	Þ.		0.400.4	~~~`	D 010	7.017	2.452	1 177	4.470	0.6270	4.627	
- HIA category +			0.4004	° ¢	Row316	7.217	3.453	1.177	4.478	-0.6378	4.627	
logP 0 -> 3.5 🖳 💶				8	1							
l hERG pIC50 ≤ 5	=		0.3941	~~~~	Row842	7.228	4.528	0.9073	4.524	-0.6858	4.424	
2D6 affinity category Iow medium 🖳 💻			0.5541	° Çi	1010-12	11220	4.520	0.5075	4.524	0.0050	4.424	
l 2C9 pKi ≤ 6 💻				~ `	1							
P-gp category no			0.3913	~O'	Row76	7.94	4.194	0.956	5.271	-0.6116	4.525	
PPB90 category Iow				\$								
BBB category -		<b>Distant</b>			1							
BBB log([brain]:[blood]) ≤ -0.5	- 5		0.3868		Row372	7.097	4.163	1.517	4.524	-0.2485	4.503	
Add rule Delete Sort Edit 📄 불 Save				54								
		linder.			1							
vailable Properties Criteria Importance	^ 6		0.3726	• • • • •	Row583	7.041	4.153	1.382	4.376	-0.4906	4.397	
hERG pIC50												
<ul> <li>HIA category</li> <li>BBB log([bra</li> </ul>		<b>History</b>		T	1							
logD	7		0.367	• <u> </u>	Row251	7.374	3.924	1.79	5.058	-0.4092	4.37	
logP				<b>*</b>								
IogS IogS @ pH7.4				$\sim$	1							
	× 8		0.3647	×	Row67	7.918	4.872	1.655	5.575	-0.4405	4.418	
coring Profiles Location				¥Y								
oral Non CNS Scoring Profile File				on the	1							
Oral CNS Scoring Profile File	9		0.359	° ch	Row421	7.094	4.412	1.537	4.857	-0.5003	4.146	
ipinski Rule of Five File ntravenous Non CNS Scoring Profile File				<u> </u>								
ntravenous CNS Scoring Profile File		lidet.		estero o	1							
lyde pKi + Oral Non CNS Scoring Project	1		0.3527	° C	Row264	7.155	3.935	1.955	4.683	-0.4545	4.33	
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O Explorer:	<			$\sim$	1							<b>`</b>

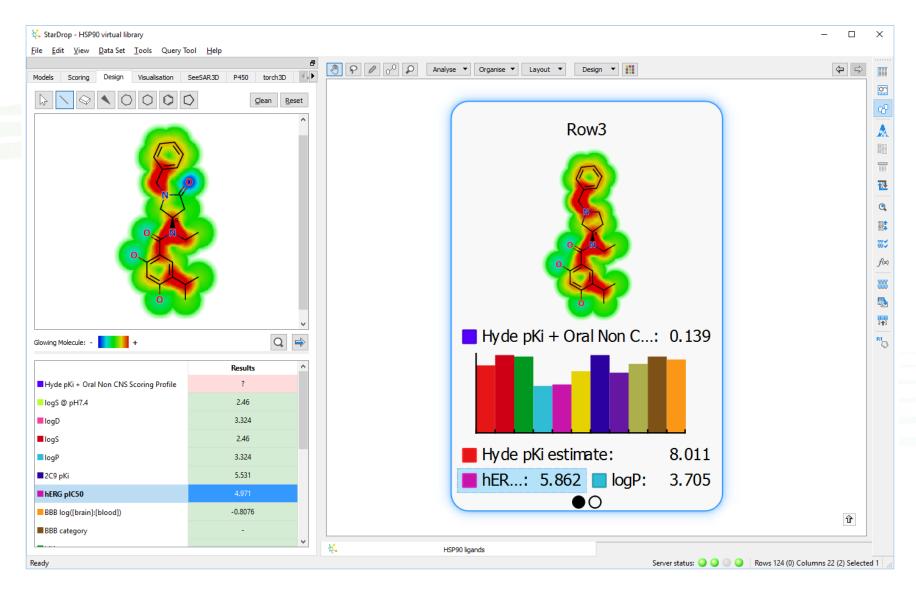
## 3D View... Optimisation opportunities



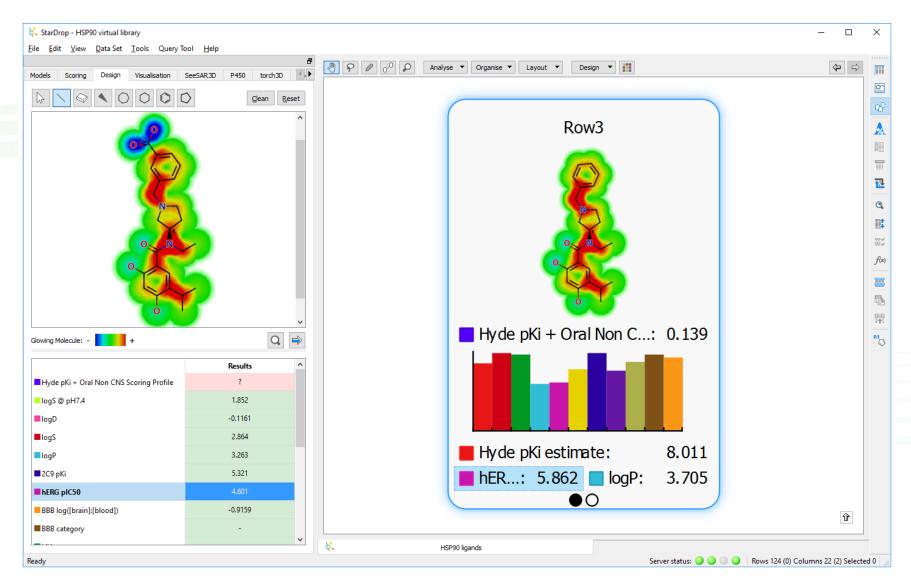
## HYDE Analysis in SeeSAR



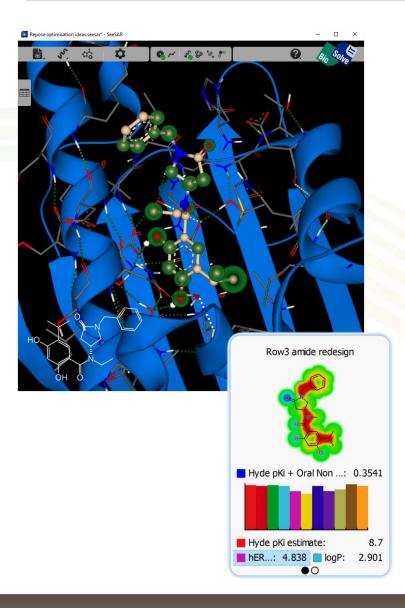
### **Optimisation Idea** Modify pKa of Nitrogen by amide substitution

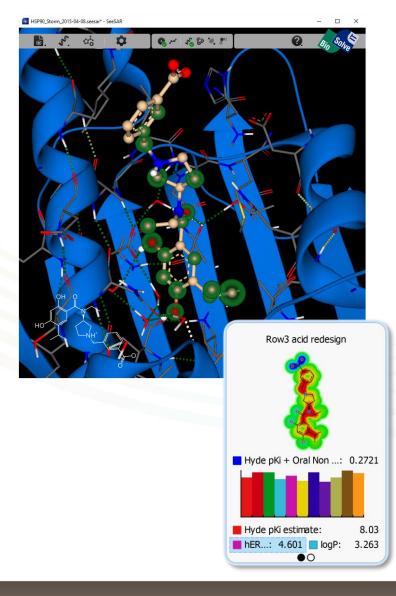


### Optimisation Idea Add polar group to phenyl ring



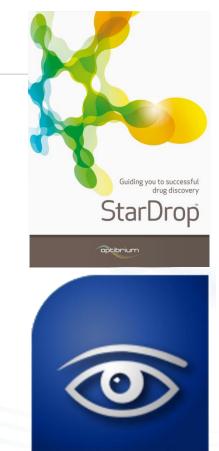
## Repose in SeeSAR





## Conclusions

- Both 2D and 3D information are important to interpret SAR and guide design
- A seamless combination between these two views of the chemical world maximises the benefits that they bring
  - Understanding SAR from experimental data
  - Analysis of virtual screening/docking results
  - Multi-parameter optimisation of potency, physicochemical and ADMET properties
- For more information:
  - www.optibrium.com/stardrop and www.biosolveit.com/SeeSAR
  - Optibrium: Booth 1227 or outside of room 6E (MEDI)

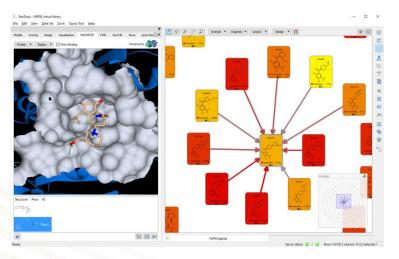


# Free Hands-On Workshop



## Seamless Integration of 2D and 3D SAR to Guide MPO

- Where: San Diego Convention Center, Room 15B
- When: Monday 3:30pm to 6pm
- Practical examples with SeeSAR and StarDrop
- Spaces are limited, so please register at Booth #1227 in the Exhibition Hall



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